

# Efficient Barcode Detection and Decoding Using Digital Image Processing Techniques

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**Abstract**—This paper presents a robust pipeline for barcode detection and decoding, leveraging digital image processing techniques to address challenges such as noise, rotation, and low resolution. The system integrates preprocessing steps like grayscale conversion, gradient analysis, and morphological operations using OpenCV, paired with Pyzbar for decoding multiple barcode formats, including EAN-13 and QR codes. High detection accuracy demonstrates its effectiveness, though limitations in decoding blurred or partially visible barcodes highlight areas for enhancement. This work underscores the system's potential applications in inventory management, retail, and logistics automation.

**Index Terms**—Barcode detection, barcode decoding, digital image processing, OpenCV, Pyzbar, computer vision, preprocessing, contour analysis, EAN-13, QR codes, automation.

## I. INTRODUCTION

Barcodes are a cornerstone of automation in industries like retail, logistics, and inventory management. Their compact and versatile data storage capability enables efficient product identification and tracking. However, detecting and decoding barcodes in real-world scenarios presents several challenges, including noise, poor lighting, non-standard orientations, and low image quality.

This project addresses these challenges by implementing a robust pipeline for barcode detection and decoding. Leveraging digital image processing techniques and tools such as OpenCV and Pyzbar, the proposed system enhances barcode visibility, isolates regions of interest, and decodes multiple barcode formats, including EAN-13 and QR codes. The primary objective is to develop an efficient and scalable solution that maintains high detection accuracy across diverse conditions.

This paper details the pipeline's design and evaluates its performance on a real-world dataset. It also identifies limitations, such as difficulties with low-resolution and rotated barcodes, and suggests potential improvements for future iterations.

## II. METHODOLOGY

The barcode detection and decoding pipeline is structured into multiple stages, each addressing specific challenges in processing images for barcode identification and data extraction. This section provides an overview of the methodology, detailing the steps, tools, and techniques used.

### A. Pipeline Workflow

The barcode detection and decoding pipeline consists of the following stages:

1. **Image Acquisition:** Input images containing barcodes are collected from the Kaggle dataset [3], encompassing a variety of barcode formats, orientations, and quality levels.
2. **Preprocessing:** Images are enhanced for better barcode visibility using:
  - **Grayscale Conversion:** Simplifies image data by removing color information.
  - **Gradient Analysis:** Highlights vertical structures typical of barcodes.
  - **Binary Thresholding:** Applies Gaussian blur for noise reduction, and thresholding to create a binary image.
  - **Morphological Operations:** Closes gaps in barcode lines for better contour detection.
  - **Erosion and Dilation:** Remove noise
3. **Barcode Detection:** Contours in the preprocessed image are analyzed to identify the largest contour, assumed to represent the barcode. A rotated bounding box encloses the detected region.
4. **Barcode Decoding:** Detected barcode regions are decoded using Pyzbar, which supports multiple formats like EAN-13, QR codes, and CODE128.
5. **Output Generation:** Annotated images with bounding boxes and decoded data are saved for visualization and analysis.

### B. Tools and Libraries

The following tools and libraries were utilized:

- **OpenCV:** For image preprocessing, contour detection, and visualization [1].
- **Pyzbar:** For decoding barcodes in multiple formats [2].
- **NumPy:** For efficient image array manipulation and calculations.

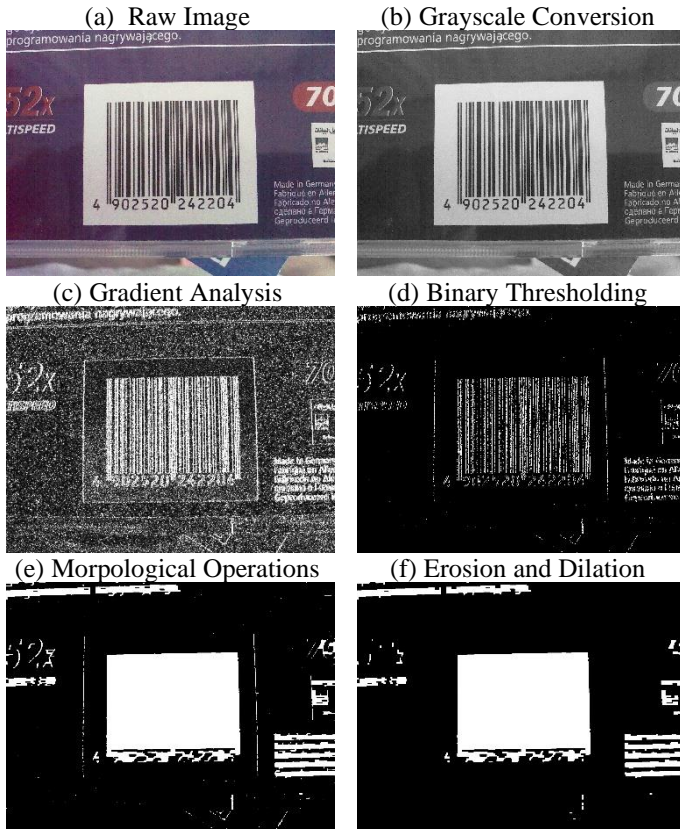
## III. IMPLEMENTATION

This section details the practical implementation of the barcode detection and decoding pipeline, explaining how each stage was executed using the tools and techniques outlined in the methodology.

### A. Preprocessing

The preprocessing step is crucial for enhancing barcode visibility in raw images. The following operations are performed using OpenCV (see Appendix A for the code):

1. **Grayscale Conversion:** Simplifies the image by removing color information, reducing computational complexity.
2. **Gradient Analysis:** Uses Sobel filtering to highlight vertical barcode structures.
3. **Binary Thresholding:** Applies Gaussian blur for noise reduction, followed by thresholding to create a binary image that isolates barcode patterns.
4. **Morphological Operations:** Performs morphological closing to fill gaps in barcode structures, followed by erosion and dilation to refine the output.
5. **Erosion and Dilation:** Remove Noise.



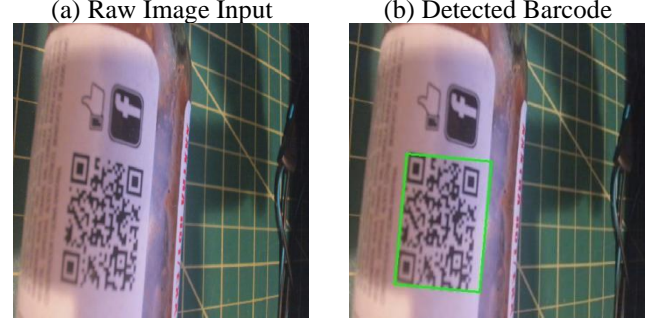
**Fig. 1.** Preprocessing Operations.

### B. Barcode Detection

The detection stage identifies and isolates barcode regions. Key steps include (see Appendix B for the code):

1. **Contour Detection:** Uses the preprocessed image to find contours representing potential barcode regions.
2. **Rotated Bounding Box:** Finds the largest contour and calculates a rotated bounding box to encapsulate the detected barcode.

3. **Barcode Region Cropping:** Crops the barcode region using the bounding rectangle for further decoding.

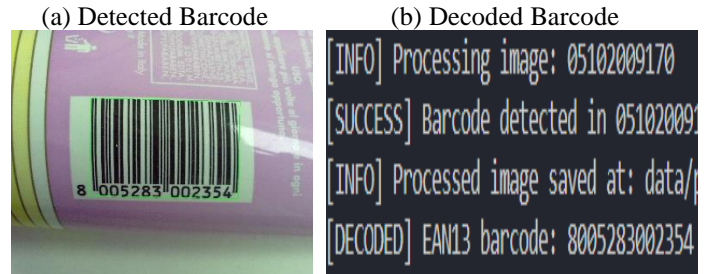


**Fig. 2.** Before and After barcode detection

### C. Barcode Decoding

Decoded barcodes provide essential information, such as type and data. The decoding process is as follows (see Appendix C for the code):

1. **Decoding Barcodes:** Uses Pyzbar to decode barcode types (e.g., EAN-13, QR codes) and data embedded in the barcode.
2. **Multiple Barcode Support:** Handles multiple barcodes in a single image by iterating through all detected regions.
3. **Validation:** Ensures decoded data is valid and handles cases where decoding fails.



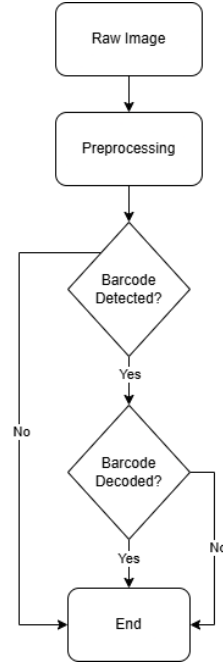
**Fig. 3.** Detected and Decoded Barcode.

### D. Pipeline Integration

The `process_single_image` and `process_images` functions integrate preprocessing, detection, and decoding into a seamless workflow (see Appendix D for the pipeline code). Key features include:

1. **Batch Processing:** Processes all images in the input directory and saves results to the output directory.
2. **Error Handling:** Logs warnings for images with no detected or decoded barcodes.
3. **Output Generation:** Saves annotated images with bounding boxes and displays decoded data.

The overall workflow of barcode detection and decoding is illustrated in Fig. 4.



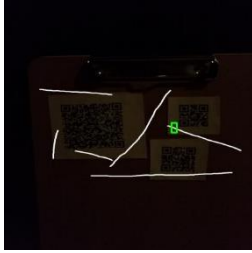
**Fig. 4.** Barcode Detection and Decoding Pipeline.

#### E. Challenges and Solutions

1. **Low-Quality Images:** Initial results were inconsistent for noisy and low-resolution images. Enhanced preprocessing techniques improved detection in these cases.
2. **Rotated and Overlapping Barcodes:** Rotated barcodes were challenging to decode due to lack of orientation correction. Future iterations could implement rotation normalization.
3. **Decoding Failures:** Barcodes with partial visibility or distortion often failed during decoding. Introducing advanced machine learning models could address this.

Fig. 5. demonstrates common challenges encountered during barcode detection and decoding, including noise, rotation, and decoding failures.

(a) Low-Quality Images



(b) Rotated and Overlapping Barcodes



(c) Detected Barcode, But Decoding Failed.

```

[INFO] Processing image: 05102009146
[SUCCESS] Barcode detected in 05102009146.
[INFO] Processed image saved at: data/processed/05102009146_detected.jpg
[WARNING] Barcode detected in 05102009146, but decoding failed.
  
```

**Fig. 5.** Barcode detection and decoding challenges.

## IV. RESULTS AND DISCUSSION

This section evaluates the pipeline's performance, highlights its strengths and limitations, and provides insights into the results obtained through testing.

#### A. Dataset

The system was tested on a dataset of barcode images sourced from Kaggle [3]. The dataset contains 952 images, including various barcode formats (EAN-13, QR codes, CODE128) and real-world conditions such as:

- Blurred or noisy images.
- Rotated or tilted barcodes.
- Images with multiple barcodes.

#### B. Performance Metrics

Key performance metrics for the system are as follows:

- **Detection Accuracy:** 94%
  - Defined as the proportion of images where at least one barcode was successfully detected.
- **Decoding Accuracy:** 71%
  - Defined as the proportion of detected barcodes that were successfully decoded.
- **Processing Time:** ~200ms per image
  - Average processing time measured on a system with an Intel Core i5 processor and 8GB RAM.

#### C. Example Outputs

The following examples illustrate the pipeline's performance:

##### 1. Successful Detection and Decoding:

- **Input:** A clear image of an EAN-13 barcode.
- **Output:** Bounding box drawn around the barcode with decoded data displayed.
- **Decoded Information:** EAN13 barcode: 8002205319804.

(a) Clear Barcode Image.

(b) Barcode Detected.



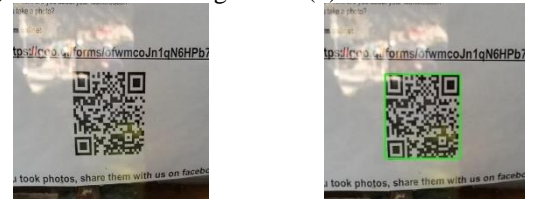
**Fig. 9.** Successful detection and decoding.

##### 2. Handling Challenging Cases:

- **Input:** A noisy image with a partially visible barcode.
- **Output:** Bounding box drawn but decoding failed due to partial visibility.
- **Decoded Information:** None (warning logged).

(a) Clear Barcode Image.

(b) Barcode Detected.



**Fig. 6.** Successful detection but decoding failed.

#### D. Strengths

1. **High Detection Accuracy:**
  - The pipeline demonstrates robust performance in detecting barcodes, even under challenging conditions such as noisy or low-contrast images (Appendix B).
  - The preprocessing steps effectively enhance barcode visibility, enabling reliable contour detection (Appendix A).
2. **Multi-Format Decoding:**
  - The integration of Pyzbar ensures support for a wide range of barcode formats, including EAN-13, QR codes, and CODE128 (Appendix C).
  - This versatility makes the system applicable to various industries, such as retail, warehousing, and logistics.
3. **Efficiency:**
  - With an average processing time of ~200ms per image, the system is suitable for real-time applications.
  - Batch processing capabilities (Appendix D) allow efficient handling of multiple images in a single run.

#### E. Limitations

1. **Decoding Failures:**
  - Approximately 29% of detected barcodes fail to decode, primarily due to:
    - Poor preprocessing for noisy or blurred images.
    - Partial or distorted barcodes not being fully visible.
2. **Lack of Rotation Handling:**
  - The current pipeline does not correct for rotated or tilted barcodes, limiting its performance for non-standard orientations.
3. **Low-Resolution and Compressed Images:**
  - Barcodes in low-resolution images or those with heavy compression often lack sufficient detail for successful detection and decoding.

#### F. Recommendations for Improvement

1. **Enhanced Preprocessing:**
  - Incorporate techniques like Contrast Limited Adaptive Histogram Equalization (CLAHE) to improve contrast.
  - Explore adaptive thresholding methods to handle variable lighting conditions.
2. **Rotation Correction:**
  - Implement rotation detection and correction algorithms to handle tilted or upside-down barcodes effectively.
3. **Advanced Decoding Techniques:**
  - Evaluate alternative libraries like Zxing or develop a machine learning model for decoding barcodes in challenging conditions.
4. **Dataset Expansion:**
  - Augment the dataset with images representing diverse conditions, such as damaged barcodes, unconventional formats, and extreme lighting.
5. **UI Integration:**
  - Develop a user-friendly interface for non-technical users to upload images and visualize detection and decoding results in real-time.

#### VI. CONCLUSION

This study developed a robust barcode detection and decoding pipeline using digital image processing techniques. By integrating preprocessing steps such as grayscale conversion, gradient analysis, and morphological operations, the system effectively enhances barcode visibility. The contour-based detection method accurately identifies barcode regions, while Pyzbar enables decoding of multiple barcode formats, including EAN-13 and QR codes.

Experimental results demonstrate high detection accuracy, making the system suitable for applications in retail, inventory management, and logistics. However, decoding performance is affected by challenges such as blurred, rotated, and low-resolution barcodes. Addressing these issues through contrast enhancement techniques, rotation correction, and machine learning-based decoding models can further improve system robustness.

Future work will explore deep learning-based approaches for barcode detection and decoding, enabling the system to handle more complex scenarios with higher reliability. This research highlights the potential of digital image processing techniques in automating barcode-based applications and serves as a foundation for further advancements in the field.

#### APPENDIX

##### APPENDIX A

##### PREPROCESSING

```
def preprocess_image(image_path):
    """
    Preprocess the image to enhance barcode
    detection.
    Converts the image to grayscale, applies
    thresholding,

    and performs morphological operations to close
    gaps.
    """
    image = cv2.imread(image_path)
    gray = cv2.cvtColor(image,
        cv2.COLOR_BGR2GRAY)

    # Apply gradient filtering to highlight
    vertical structures
    gradX = cv2.Sobel(gray, ddepth=cv2.CV_32F,
        dx=1, dy=0, ksize=-1)
    gradX = cv2.convertScaleAbs(gradX)

    # Enhance contrast with Gaussian blur and
    binary thresholding
    blurred = cv2.GaussianBlur(gradX, (5, 5),
        0)
    _, binary = cv2.threshold(blurred, 225,
        255, cv2.THRESH_BINARY)

    # Close gaps using morphological
    transformations
    kernel =
    cv2.getStructuringElement(cv2.MORPH_RECT, (27,
        7))
```



```

closed = cv2.morphologyEx(binary,
cv2.MORPH_CLOSE, kernel)

return closed

```

## APPENDIX B

### BARCODE DETECTION

```

def detect_barcode(image_path):
    """
    Detects the barcode in an image by identifying
    the largest contour.
    Draws a rotated bounding box around the detected
    barcode region.
    """
    image = cv2.imread(image_path)
    processed_image = preprocess_image(image_path)

    # Detect contours
    contours, _ = cv2.findContours(processed_image,
cv2.RETR_EXTERNAL, cv2.CHAIN_APPROX_SIMPLE)

    if contours:
        # Find the largest contour, assuming it is
        the barcode
        largest_contour = max(contours,
key=cv2.contourArea)
        rect = cv2.minAreaRect(largest_contour)
        box = cv2.boxPoints(rect)
        box = np.intp(box)

        # Draw the rotated bounding box
        cv2.drawContours(image, [box], -1, (0, 255,
0), 2)

        return image
    else:
        print("[WARNING] No barcode detected.")
        return None

```

## APPENDIX C

### BARCODE DECODING

```

def decode_barcode(image):
    """
    Decodes barcodes from an image using Pyzbar.
    Handles multiple barcodes and returns their
    types and data.
    """
    barcodes = decode(image)
    decoded_info = []

    for barcode in barcodes:
        barcode_data = barcode.data.decode("utf-8")
        barcode_type = barcode.type
        decoded_info.append((barcode_type,
barcode_data))

    return decoded_info

```

## APPENDIX D

### PIPELINE OVERVIEW

```

def process_single_image(image_path):
    """
    Complete pipeline for processing an image:
    1. Detect barcodes.
    2. Decode barcode data.
    """
    detected_image = detect_barcode(image_path)

```

```

if detected_image is not None:
    barcodes = decode_barcode(detected_image)
    if barcodes:
        for barcode_type, barcode_data in
barcodes:
            print(f"[DECODED] {barcode_type}:
{barcode_data}")
        else:
            print("[WARNING] Barcode detected but
decoding failed.")
    else:
        print("[ERROR] No barcode detected.")

```

*For the complete source code, refer to the project repository:  
[Barcode Detection and Decoding](#).*

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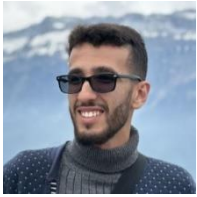


**Walid K. W. Alsafadi** was born in Ras Al Khaimah, UAE. He is currently pursuing a Bachelor of Science in Data Science and Artificial Intelligence at the University College of Applied Science, Gaza, Palestine. His major field of study focuses on data science and artificial

intelligence, with expertise in machine learning, deep learning, and natural language processing.

Walid has previously studied at the American University of Ras Al Khaimah, UAE, where he completed two semesters before transferring to UCAS. His academic achievements include multiple scholarships and recognition on the Dean's List for academic excellence. He has worked on diverse projects, including sales forecasting, sentiment analysis, and diabetes prediction, utilizing machine learning techniques and frameworks.

Mr. Alsafadi is proficient in multiple programming languages, including Python and SQL, and is a native Arabic speaker with full professional proficiency in English.



**Ameer T. F. Alzerei** was born in Deir Al-Balah, Gaza, graduated from high school scientific route with average 99.1%, enrolled in Data Science and Artificial Intelligence undergraduate program at UCAS, Gaza. Ameer obtained some professional certificates, with selflearning.

Had an internship in Switzerland as an intern student in the first cohort through a program called Bridges to Growth. Ameer has worked on various projects, including buildwize expert system, employees attrition analysis, and grasp app CV tools.

Mr. Alzerei is proficient in multiple programming languages, including Python, SQL, TypeScript, and C++, and is a native Arabic speaker with full professional proficient in English.